Integer-Forcing

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Classical Decoding Strategies:

- Joint Decoding
- Zero-Forcing
- Successive Interference Cancellation

A Simple Example

•
$$\mathbf{Y} = \begin{bmatrix} 1 & 2 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{bmatrix} + \mathbf{Z}$$

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- Zero-Forcing: $\begin{bmatrix} -1 & 2\\ 1 & -1 \end{bmatrix} \mathbf{Y} = \begin{bmatrix} \mathbf{x}_1\\ \mathbf{x}_2 \end{bmatrix} + \begin{bmatrix} -1 & 2\\ 1 & -1 \end{bmatrix} \mathbf{Z}$
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- Effective noise variances: $\sigma_1^2 = 5$ and $\sigma_2^2 = 2$.
- Integer-Forcing: $\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \mathbf{Y} = \begin{bmatrix} \mathbf{x}_1 + 2\mathbf{x}_2 \\ \mathbf{x}_1 + \mathbf{x}_2 \end{bmatrix} + \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \mathbf{Z}$
- Effective noise variances: $\sigma_1^2 = 1$ and $\sigma_2^2 = 1$.



- Possible to decode linear functions of the transmitted messages. (Nazer-Gastpar '11)
- Provided that the codebooks share a common algebraic structure.

Compute-and-Forward



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Compute-and-Forward: Fading Channels



- Finite field messages, $\mathbf{w}_{\ell} \in \mathbb{F}_q^k$.
- Gaussian channel model, $\|\mathbf{x}_{\ell}\|^2 \leq nP$ and $\mathbf{z}_m \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$.

• Equal rates
$$R = \frac{k}{n} \log_2 q$$
.

- Each receiver wants a linear combination, $\mathbf{u}_m = \bigoplus a_{m\ell} \mathbf{w}_\ell$
- Vanishing probability of error, $\lim_{n\to\infty} \mathbb{P}(\bigcup_m \{ \hat{\mathbf{u}}_m \neq \mathbf{u}_m \}) = 0.$

M

 $\ell = 1$

Effective Noise

$$\mathbf{y}_{m} = \sum_{\ell=1}^{M} h_{m\ell} \mathbf{x}_{\ell} + \mathbf{z}_{m}$$
$$= \sum_{\ell=1}^{M} a_{m\ell} \mathbf{x}_{\ell} + \sum_{\ell=1}^{M} (h_{m\ell} - a_{m\ell}) \mathbf{x}_{\ell} + \mathbf{z}_{m}$$
Effective Noise

- How can we go between the integer combination of the real-valued codewords and the linear combination of the finite field messages?
- How do we cope with the self-noise?
- Use (dithered) nested lattice codes from Erez-Zamir '04.

Nested Lattice Codes from q-ary Linear Codes

• Choose an $n \times k$ generator matrix $\mathbf{G} \in \mathbb{F}_q^{n \times k}$ for q-ary code.



- Map elements $\{0, 1, 2, \dots, q-1\}$ to equally spaced points between -1/2 and 1/2.
- Place codewords $\mathbf{x} = \mathbf{G}\mathbf{w}$ into 0 the fundamental Voronoi region $(-\frac{1}{2}, -\frac{1}{2})$ $\mathcal{V} = [-1/2, 1/2)^n$



• One more step needed for shaping gain...

Effective Noise

- Only need channel state information at the receivers.
- Effective noise due to mismatch between channel coefficients $\mathbf{h}_m = [h_{m1} \cdots h_{mM}]^T$ and equation coefficients $\mathbf{a}_m = [a_{m1} \cdots a_{mM}]^T$.

$$N_{\mathsf{EFFEC},m} = 1 + P \|\mathbf{h}_m - \mathbf{a}_m\|^2$$
$$R = \min_m \frac{1}{2} \log^+ \left(\frac{P}{1 + P \|\mathbf{h}_m - \mathbf{a}_m\|^2}\right)$$

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• Can do better with MMSE scaling.

$$\begin{split} N_{\mathsf{EFFEC},m} &= \alpha_m^2 + P \|\alpha_m \mathbf{h}_m - \mathbf{a}_m\|^2 \\ R &= \min_m \max_{\alpha_m} \frac{1}{2} \log \left(\frac{P}{\alpha_m^2 + P \|\alpha_m \mathbf{h}_m - \mathbf{a}_m\|^2} \right) \\ &= \min_m \frac{1}{2} \log^+ \left(\frac{1 + P \|\mathbf{h}_m\|^2}{\|\mathbf{a}_m\|^2 + P(\|\mathbf{h}_m\|^2 \|\mathbf{a}_m\|^2 - (\mathbf{h}_m^T \mathbf{a}_m)^2)} \right) \end{split}$$



• M transmit and N receive antennas ($N \ge M$).

- Channel model: $\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{Z}$
- Each antenna encodes an independent message w_ℓ (or data stream) of rate R/M to a codeword x_ℓ ∈ ℝ^T with power at most SNR.
- i.i.d. Gaussian noise $\mathbf{z}_m \sim \mathcal{N}(\mathbf{0}, \mathbf{I})$.
- Probability of error: $\mathbb{P}({\{\hat{\mathbf{w}}_1 \neq \mathbf{w}_1\} \cup \cdots \cup \{\hat{\mathbf{w}}_M \neq \mathbf{w}_M\}}) < \epsilon$
- V-BLAST setting.



· Joint maximum likelihood decoder offers optimal performance,

$$R_{\mathsf{JOINT}}(\mathbf{H}) = \min_{\mathcal{S} \subseteq \{1, 2, \dots, M\}} \frac{M}{2|\mathcal{S}|} \log \det \left(\mathbf{I}_{\mathcal{S}} + \mathsf{SNR} \ \mathbf{H}_{\mathcal{S}} \mathbf{H}_{\mathcal{S}}^T \right)$$

- Worst-case complexity is exponential in MT (where M =number of antennas and T =blocklength).
- We would like to approach this performance as closely as possible using single stream decoding.

Linear Receiver Architectures



- First project the received signal, $\tilde{\mathbf{Y}} = \mathbf{B}\mathbf{Y}$.
- Each projected output stream $\tilde{\mathbf{y}}_m$ is fed into a decoder that attempts to recover a single message \mathbf{w}_m .
- Each decoder may face interference from other data streams,

$$R_{\mathsf{LINEAR}}(\mathbf{H}) = \min_{m} \frac{M}{2} \log \left(1 + \frac{\mathsf{SNR} \|\mathbf{b}_{m}^{T} \mathbf{h}_{m}\|^{2}}{\|\mathbf{b}_{m}\|^{2} + \mathsf{SNR} \sum_{i \neq m} \|\mathbf{b}_{m}^{T} \mathbf{h}_{i}\|^{2}} \right)$$

• Worst-case complexity is exponential in T and linear in M.

• Zero-Forcing Receiver: Eliminate interference between data streams by setting projection matrix to

$$\mathbf{B} = \mathbf{H}^{\dagger}$$

- Also known as a decorrelator.
- Effective channel: $\tilde{\mathbf{Y}} = \mathbf{X} + \mathbf{H}^{-1}\mathbf{Z}$ (if \mathbf{H} is full rank). Performs poorly if \mathbf{H} is ill-conditioned.

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- MMSE Receiver: Can do slightly better with a regularized projection matrix

$$\mathbf{B} = \mathbf{H}^T \left(\mathbf{H} \mathbf{H}^T + \frac{1}{\mathsf{SNR}} \mathbf{I} \right)^{-1}$$

Integer-Forcing Receiver Architectures



- Each stream $\tilde{\mathbf{y}}_m$ is fed into a decoder that attempts to recover an equation \mathbf{u}_m .
- Equations can then be solved digitally for the original messages.
- Choose equation coefficients A to minimize effective noise.
- Worst-case complexity is exponential in T and exponential in M.

Integer-Forcing Receiver Architectures



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• Zhan-Nazer-Erez-Gastpar, submitted to IT <arXiv:1003.5966> The achievable rate of integer-forcing is

$$R_{\mathsf{INTEGER}}(\mathbf{H}) = \max_{\substack{\mathbf{A} \in \mathbb{Z}^{M \times M} \\ \mathsf{rank}(\mathbf{A}) = M}} \max_{\mathbf{B} \in \mathbb{R}^{M \times N}} \min_{m} R(\mathbf{H}, \mathbf{a}_{m}, \mathbf{b}_{m})$$
$$R(\mathbf{H}, \mathbf{a}_{m}, \mathbf{b}_{m}) = \frac{M}{2} \log^{+} \left(\frac{\mathsf{SNR}}{\|\mathbf{b}_{m}\|^{2} + \mathsf{SNR}\|\mathbf{H}^{T}\mathbf{b}_{m} - \mathbf{a}_{m}\|^{2}} \right)$$

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• Only need to search over integer vectors \mathbf{a}_m satisfying

$$\|\mathbf{a}_m\|^2 \le 1 + \lambda_{\mathsf{MAX}}^2 \mathsf{SNR}$$

where λ_{MAX} is the maximum singular value of $\mathbf{H}.$

• Exact Integer-Forcing: Eliminate interference between data streams by setting projection matrix to

$$\mathbf{B} = \mathbf{A}\mathbf{H}^{\dagger}$$
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• MMSE Integer-Forcing: Can do slightly better with a regularized projection matrix

$$\mathbf{B} = \mathbf{A}\mathbf{H}^T \left(\frac{1}{\mathsf{SNR}}\mathbf{I} + \mathbf{H}\mathbf{H}^T\right)^{-1}$$

- Channel matrix \mathbf{H} is i.i.d. Gaussian, only known at the receiver.
- Assume there exists an architecture that encodes each data stream at the same rate and achieves sum rate $R(\mathbf{H})$. For a target rate R, then the outage probability is defined as

$$p_{\mathsf{OUT}}(R) = \Pr(R(\mathbf{H}) < R).$$

• For a fixed probability $p \in (0,1]$, the *outage rate* is defined to be

$$R_{\mathsf{OUT}}(p) = \sup\{R : p_{\mathsf{OUT}}(R) \le p\}.$$

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- V-BLAST II: Select the decoding order for each channel realization to maximize the effective SNR for the data stream that sees the worst channel.
- V-BLAST III: Decodes and cancel the data streams in a predetermined order. The rate of each data stream is selected to maximize the sum rate. (Outside problem statement.)

Simulation: Outage Rates



Figure: 1 percent outage rates for the 2×2 complex-valued MIMO channel with Rayleigh fading.



Figure: Outage probability for the 2×2 complex-valued MIMO channel with Rayleigh fading for a target sum rate of R = 6

$$\widetilde{\sigma}_m^2 = |\mathbf{v}_{\min}^T \mathbf{a}_m|^2 \frac{1}{\lambda_{\min}} + |\mathbf{v}_{\max}^T \mathbf{a}_m|^2 \frac{1}{\lambda_{\max}}$$

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• Zheng-Tse '03: A family of codes is said to achieve spatial multiplexing gain r and diversity gain d if the total data rate and the average probability of error satisfy

$$\begin{split} \lim_{\mathsf{SNR}\to\infty} \frac{R(\mathsf{SNR})}{\log\mathsf{SNR}} \geq r\\ \lim_{\mathsf{SNR}\to\infty} \frac{\log P_e(\mathsf{SNR})}{\log\mathsf{SNR}} \leq -d. \end{split}$$

• For our problem, the optimal diversity-multiplexing tradeoff (DMT) under Rayleigh fading is

$$d_{\rm joint}(r) = N\left(1 - \frac{r}{M}\right)$$

where $r \in [0, M]$ and can be achieved by joint ML decoding.

• Zheng-Tse '03: The DMTs achieved by the decorrelator and SIC architectures are

$$d_{\text{decorr}}(r) = \left(1 - \frac{r}{M}\right)$$
$$d_{\text{v-blast I}}(r) = \left(1 - \frac{r}{M}\right)$$
$$d_{\text{v-blast II}}(r) \le (N - 1)\left(1 - \frac{r}{M}\right)$$

 $d_{\text{V-BLAST III}}(r) = \text{piecewise linear curve connecting points } (r_k, n-k)$

where
$$r_0 = 0, r_k = \sum_{i=0}^{k-1} \frac{k-i}{n-i} \ 1 \le k \le n$$

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• Integer-forcing recovers the optimal DMT:

$$d_{\rm integer}(r) = N\left(1 - \frac{r}{M}\right)$$



Figure: Diversity-multiplexing tradeoff for the 4×4 MIMO channel with Rayleigh fading.

MIMO with Interference



Choose integer vectors orthogonal to the interference space.



 $\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{J}\mathbf{S} + \mathbf{Z}$

Integer-Forcing Advantage



2 by 2 MIMO, 1 dimensional interference, $\mathsf{INR}=\mathsf{SNR}^{0.2}$

- Lattice Reduction: Yao-Wornell '02, Taherzadeh-Mobasher-Khandani '07, Jalden-Elia '09
- Lattices for AWGN Capacity: Erez-Zamir '04
- Lattices for DMT: El Gamal-Caire-Damen '04
- Practical compute-and-forward: Feng-Silva-Kschischang '10, Hern and Narayanan '11, Ordentlich and Erez '10, Osmane and Belfiore '11

- Algebraic structure enables us to do something in between treating interference as noise and decoding interference.
- See ISIT '11 Tutorial: Algebraic Structure in Network Information Theory for more information.
- Near-optimal receiver architecture with decoupled optimization problems.
- Question: Can this be generalized to include encoding across transmit antennas?

Even more V-BLAST

• V-BLAST I-V: Decodes and cancel the data streams in the best order, given the channel realization. The rate of each data stream is selected to maximize the sum rate. (Outside problem statement.)

